

# ECONOMETRIC ANALYSIS

Bianca INOCENCIO  
Glebs SOLOVJOVS  
Rhita SQALLI



# Possible models

1. **Fixed Effects (Within):** incorporates entity-specific variables into the model to account for unobserved variability across entities. It accounts for unobserved, time-invariant factors that could have an impact on the dependent variable. The FE estimator captures the within-entity variation over time by incorporating entity-fixed effects.
2. **Random Effects (GLS/FGLS):** takes into account both variance within and between entities. The entity-specific effects are assumed by the RE model to be uncorrelated with the independent variables, while they might be correlated with the error term. The correlation between the entity-specific effects and the error term is addressed using the GLS or FGLS estimator.
3. **Pooled OLS:** ignores entity-specific effects and time dependencies in favor of treating the panel data as a single cross-sectional dataset. Without making a distinction between different things or times, it calculates a single regression equation. Pooled OLS makes the assumption that there is no hidden heterogeneity or correlation between objects or across time.
4. **First Difference:** to address unobserved time-invariant heterogeneity in panel data analysis. It involves taking the difference between consecutive observations for each entity, effectively eliminating the time-invariant individual-specific factors from the analysis.

First, we set up our data panel with `xtset ivar tvar`, and then we look at the summary, we see that we have data for all individuals for all periods of time and so we can say that we have a **balanced panel**.

```
. xtset ivar tvar
```

```
Panel variable: ivar (strongly balanced)
Time variable: tvar, 1 to 5
Delta: 1 unit
```

```
. xtdescribe
```

```
ivar: 1, 2, ..., 604      n =      604
tvar: 1, 2, ..., 5        T =        5
Delta(tvar) = 1 unit
Span(tvar)  = 5 periods
(ivar*tvar uniquely identifies each observation)
```

```
Distribution of T_i:  min    5%   25%   50%   75%   95%   max
                   5      5      5      5      5      5
```

Freq.	Percent	Cum.	Pattern
604	100.00	100.00	11111
604	100.00		XXXXX



# FE Estimator

The three R-sq measure the degree to which the LSDV estimated model can explain the within, the between, and the overall variation of the response variable y.

```
. xtreg y k l ,fe
```

```
Fixed-effects (within) regression      Number of obs   =    3,020
Group variable: ivar                  Number of groups =     604

R-squared:                            Obs per group:
    Within = 0.1802                    min =          5
    Between = 0.8810                  avg =         5.0
    Overall = 0.5965                  max =          5

corr(u_i, Xb) = 0.6866                F(2, 2414)      =   265.24
                                      Prob > F         =   0.0000
```

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
k	.4876862	.0235142	20.74	0.000	.441576	.5337964
l	.3895744	.0390815	9.97	0.000	.3129376	.4662111
_cons	1.533569	.0559373	27.42	0.000	1.423879	1.643259
sigma_u	.54061562					
sigma_e	.56805995					
rho	.475261	(fraction of variance due to u_i)				

F test that all u\_i=0: F(603, 2414) = 1.73

Prob > F = 0.0000

We can see that the F-test yielded **Prob > F = 0.000** which means that the model is **significant and it has explanatory power**.

- The t-ratio test on parameters individually yielded  $P > |t| = 0.000$  which means that coefficients  $\beta_k$  and  $\beta_l$  are different from 0 (both positive).
- All of the coefficients are individually significant and have the right sign..

sigma\_e is the square root of the unbiased estimator for  $\sigma_{LSDV}$ .

\_cons is the arithmetic mean for all the  $\sigma_{LSDV}$ . It is sizable and different from zero, however, we do not know how these components differ across individuals.

The test at the end tells us whether we can use POLS or not and disregard the panel structure altogether. The null hypothesis is  $\alpha_1 = \alpha_2 = \alpha_3 = \dots = \alpha_n$  and in our case there is LH, so we cannot use **POLS**. We then check for the correlation of error u with y, which is positive.

```
. corr u y
(obs=3,020)
```

	u	y
u	1.0000	
y	0.8128	1.0000

# Robust FE and Two-way FE Estimator

## Robust Variance:

We can see that almost 60% of the variation is described by the model, while standard error is comparable to the regular FE. All the coefficients are significant.

```
. xtreg y k l ,fe vce(robust)
```

Fixed-effects (within) regression  
Group variable: ivar

Number of obs = 3,020  
Number of groups = 604

R-squared:  
Within = 0.1802  
Between = 0.8810  
Overall = 0.5965

Obs per group:  
min = 5  
avg = 5.0  
max = 5

corr(u\_i, Xb) = 0.6866  
F(2, 603) = 254.23  
Prob > F = 0.0000

(Std. err. adjusted for 604 clusters in ivar)

y	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
k	.4876862	.0246763	19.76	0.000	.4392243	.5361481
l	.3895744	.0377622	10.32	0.000	.315413	.4637357
_cons	1.533569	.0531233	28.87	0.000	1.42924	1.637898
sigma_u	.54061562					
sigma_e	.56805995					
rho	.475261				(fraction of variance due to u_i)	

## Two-way estimator:

```
. quietly tabulate tvar,gen(time_d)
```

```
. xtreg y k l time_d1-time_d5 ,fe  
note: time_d5 omitted because of collinearity.
```

Fixed-effects (within) regression  
Group variable: ivar

Number of obs = 3,020  
Number of groups = 604

R-squared:  
Within = 0.3619  
Between = 0.8823  
Overall = 0.6203

Obs per group:  
min = 5  
avg = 5.0  
max = 5

corr(u\_i, Xb) = 0.5951

F(6, 2410) = 227.76  
Prob > F = 0.0000

y	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
k	.4838285	.0208046	23.26	0.000	.4430317	.5246252
l	.3954738	.0345223	11.46	0.000	.3277775	.4631702
time_d1	-.128353	.0289162	-4.44	0.000	-.1850561	-.0716499
time_d2	-.2735202	.0288738	-9.47	0.000	-.3301403	-.2169001
time_d3	.2459541	.0288943	8.51	0.000	.1892939	.3026144
time_d4	.3795736	.0288825	13.14	0.000	.3229366	.4362107
time_d5	0 (omitted)					
_cons	1.483542	.0524324	28.29	0.000	1.380725	1.586359
sigma_u	.53877886					
sigma_e	.5015906					
rho	.53569962				(fraction of variance due to u_i)	

F test that all u\_i=0: F(603, 2410) = 2.21 Prob > F = 0.0000

```
. testparm time_d1-time_d5
```

```
(1) time_d1 = 0  
(2) time_d2 = 0  
(3) time_d3 = 0  
(4) time_d4 = 0
```

F( 4, 2410) = 171.55  
Prob > F = 0.0000

When using the two-way FE the model explains the variation slightly better. From *testparm* we see that there is no homogeneity as the coefficients are different from each other. The model overall and all coefficients are significant.



# RE Estimator

The coefficients on  $l$  and  $k$  are individually significant ( $P > t = 0.000$  for both), collectively distinct from zero (Wald  $\chi^2(2) = 5149.06$ ), and positive (0.6023521, 1.067278). Then, based on this output,  $l$  and  $k$  have a highly favorable effect on  $y$ . The assumption that  $\text{corr}(u_i, Xb) = 0$ —strict exogeneity of the regressors with respect to individual effects—underlies the random effect model.

```
. xtreg y k l ,re
```

Random-effects GLS regression  
Group variable:  $ivar$

Number of obs = 3,020  
Number of groups = 604

R-squared:

Within = 0.1568  
Between = 0.9046  
Overall = 0.6387

Obs per group:

min = 5  
avg = 5.0  
max = 5

$\text{corr}(u_i, X) = 0$  (assumed)

Wald  $\chi^2(2) = 5149.06$   
Prob >  $\chi^2 = 0.0000$

	y	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
	k	.6023521	.0217996	27.63	0.000	.5596257	.6450785
	l	1.067278	.0207724	51.38	0.000	1.026565	1.107991
	_cons	.5429267	.0287017	18.92	0.000	.4866725	.599181
	sigma_u	.05502214					
	sigma_e	.56805995					
	rho	.00929461	(fraction of variance due to $u_i$ )				

- Now we test if there are random effects in the model.

```
. xttest0
```

Breusch and Pagan Lagrangian multiplier test for random effects

$y[ivar,t] = Xb + u[ivar] + e[ivar,t]$

Estimated results:

	Var	SD = sqrt(Var)
y	1.021903	1.010892
e	.3226921	.5680599
u	.0030274	.0550221

Test:  $\text{Var}(u) = 0$

$\text{chibar2}(01) = 0.28$   
Prob >  $\text{chibar2} = 0.2982$

↳ We do not reject the null hypothesis with  $\text{Prob} > \text{chibar2} = 0.2982$ , we therefore better use the FE model.

```
. regress y l k ,vce(cluster ivar)
```

Linear regression

Number of obs = 3,020  
F(2, 603) = 2374.54  
Prob > F = 0.0000  
R-squared = 0.6387  
Root MSE = .60786

(Std. err. adjusted for 604 clusters in  $ivar$ )

	y	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
	l	1.073524	.021853	49.12	0.000	1.030607	1.116441
	k	.6033576	.0224706	26.85	0.000	.5592274	.6474878
	_cons	.5338313	.030894	17.28	0.000	.4731584	.5945042



## Two-way RE

```
. xtreg y k l ib1.tvar
```

Random-effects GLS regression  
Group variable: ivar

R-squared:

Within = 0.3187  
Between = 0.9045  
Overall = 0.6949

corr(u\_i, X) = 0 (assumed)

Number of obs = 3,020  
Number of groups = 604

Obs per group:

min = 5  
avg = 5.0  
max = 5

Wald chi2(6) = 5676.26  
Prob > chi2 = 0.0000

y	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
k	.5928659	.0199839	29.67	0.000	.5536982	.6320337
l	1.033179	.0200997	51.40	0.000	.9937842	1.072573
tvar						
2	-.151342	.0311327	-4.86	0.000	-.212361	-.0903231
3	.3655029	.0311182	11.75	0.000	.3045123	.4264934
4	.5104561	.0311199	16.40	0.000	.4494622	.5714501
5	.1313356	.031116	4.21	0.000	.0702632	.192408
_cons	.4241277	.034518	12.29	0.000	.3564738	.4917817
sigma_u	.13132871					
sigma_e	.5015906					
rho	.06415417	(fraction of variance due to u_i)				

The coefficients on l and k are individually significant ( $P > t = 0.000$  for both), collectively different from zero (Wald  $\chi^2(2) = 5676.26$ ), and positive (.5928659, 1.033179). We also notice the three R-squared measurements, and since none of them were obtained using OLS regression, we can only interpret them here as squared correlation coefficients.

- **Within = 0.3187** squared correlation of 31.87% between y and fitted values both in individual-mean deviations

- **Between = 0.9045** squared correlation of 90.45% for both group means of y and the fitted values

- **Overall = 0.6949** squared correlation between the untransformed values of y and the fitted values, which is 69.49%.

This estimator performs similarly between than the between estimator, similarly better within than the within fixed-effects estimator, and slightly better overall, according to the R-squared.



# FD estimator: Autocorrelation and heteroskedasticity

- Serial correlation:

```
. xtserial y l k
```

Wooldridge test for autocorrelation in panel data

H0: no first-order autocorrelation

F( 1, 603) = 35.974

Prob > F = 0.0000

- ↳ We can observe there is first-order autocorrelation.

- Test for heteroskedasticity in the FE setting:

```
. quietly xtreg y l k ,fe
```

```
. xttest3
```

Modified Wald test for groupwise heteroskedasticity  
in fixed effect regression model

H0:  $\sigma(i)^2 = \sigma^2$  for all  $i$

chi2 (604) = 26871.23

Prob>chi2 = 0.0000

- ↳ Strong evidence against the null hypothesis of homoskedasticity, indicating the presence of groupwise heteroskedasticity. We need to address the issue by using the robust variance.

- We then look at the FD estimator:

```
. xtserial y l k ,output
```

Linear regression

Number of obs	=	2,416
F(2, 603)	=	205.87
Prob > F	=	0.0000
R-squared	=	0.1952
Root MSE	=	.77132

(Std. err. adjusted for 604 clusters in ivar)

		Robust		t	P> t	[95% conf. interval]	
D.y	Coefficient	std. err.					
1							
D1.	.3624337	.042283	8.57	0.000	.2793938	.4454735	
k							
D1.	.5057574	.0269717	18.75	0.000	.4527874	.5587273	

Wooldridge test for autocorrelation in panel data

H0: no first-order autocorrelation

F( 1, 603) = 35.974

Prob > F = 0.0000

- ↳ Once again, we see that there is first-order correlation.





# Hausman Test

- We apply the Hausman test to compare the Fixed Effects and the Random Effects models:

↳ We can observe that the null hypothesis is rejected, meaning that the random effects (RE) estimator and the fixed effects (FE) estimator are consistent. We can then infer that the Fixed Effect model is better suited for our study than the Random Effect.

```
. quietly xtreg y l k ,fe
. estimates store fe
. quietly xtreg y l k ,re
. estimates store re
. hausman fe re ,sigmaless
```

	Coefficients		(b-B) Difference	sqrt(diag(V_b-V_B)) Std. err.
	(b) fe	(B) re		
1	.3895744	1.067278	-.6777034	.0338676
k	.4876862	.6023521	-.1146659	.0115769

b = Consistent under H0 and Ha; obtained from xtreg.  
B = Inconsistent under Ha, efficient under H0; obtained from xtreg.

Test of H0: Difference in coefficients not systematic

```
chi2(2) = (b-B)'[(V_b-V_B)^(-1)](b-B)
          = 407.69
Prob > chi2 = 0.0000
```





# Robust Hausman Test

- First, we do the regression with group means.

```
. foreach x of varlist l k {
2. bysort ivar: egen gm`x'=mean(`x')
3. }

. xtreg y l k gml gmk, re vce(cluster ivar)
```

Random-effects GLS regression  
Group variable: ivar

Number of obs = 3,020  
Number of groups = 604

R-squared:  
Within = 0.1802  
Between = 0.9049  
Overall = 0.6817

Obs per group:  
min = 5  
avg = 5.0  
max = 5

corr(u\_i, X) = 0 (assumed)

Wald chi2(4) = 6285.55  
Prob > chi2 = 0.0000

(Std. err. adjusted for 604 clusters in ivar)

	y	Coefficient	Robust std. err.	z	P> z	[95% conf. interval]	
	l	.3895744	.0377747	10.31	0.000	.3155373	.4636114
	k	.4876862	.0246845	19.76	0.000	.4393055	.5360668
	gml	.8534943	.0487775	17.50	0.000	.7578921	.9490965
	gmk	.1374399	.0509745	2.70	0.007	.0375316	.2373482
	_cons	.290738	.0310231	9.37	0.000	.2299338	.3515422
	sigma_u	.05502214					
	sigma_e	.56805995					
	rho	.00929461	(fraction of variance due to u_i)				

- Now we do the test:

```
. testparm gmk gml
```

```
( 1) gml = 0
( 2) gmk = 0
```

```
chi2( 2) = 415.36
Prob > chi2 = 0.0000
```

↳ The Robust Hausman test leads us to the same conclusion than the standard Hausman test: the p-value is 0, we can reject the null hypothesis that the RE estimator and the FE estimator are consistent and choose the FE model, even after accounting for heteroskedasticity and/or serial correlation.

# Conclusion



Comparing the four specifications (Fixed effects, Random effects, Pooled OLS and First Difference) we can conclude that the Fixed Effects with robust variance is the best model to estimate the production function with panel data on  $y$  = output,  $x$  = capital; labor.

We eliminate the POLS as from the output of the FE regression we see that there is heterogeneity and so we cannot get rid of the panel structure.

Looking at the Wooldridge test for first-order autocorrelation we can see that it is present as  $H_0$  is rejected. Thus, we look at the robust Hausman test that rejects the  $H_0$  indicating that FE is more effective than RE in this situation. This is also supported by the test for random effects which indicates that they are absent from the model.

We eliminate the FD model as it requires the heterogeneity to be time-invariant, however, in our case it varies with time. This can be understood from the Hausman test that indicates that FE is better than RE and thus the heterogeneity varies with time.

As there is autocorrelation we also eliminate two-way FE as it assumes no correlation between error terms.

Lastly, to address the issue of heteroskedasticity and autocorrelation we opt for the robust variance version of FE.

The model shows us that there is (obviously) a positive relationship between labor and output, but that this relationship is less strong than the one between capital and output, which is also positive. In addition to that, we can say that the most accurate specification is the **Fixed Effects (Within) with robust variance**.